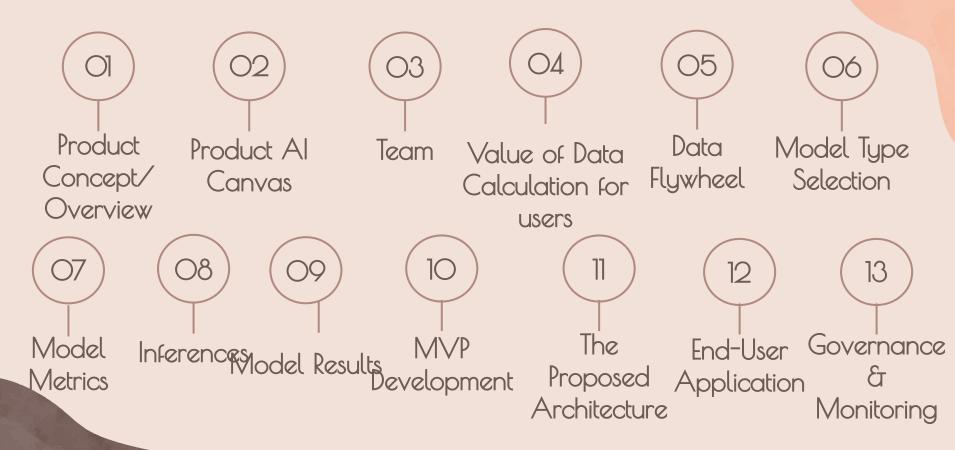
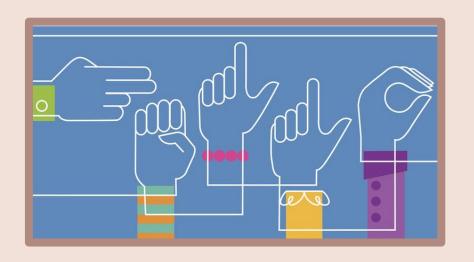


Table of content



Product Concept/Overview



To build an interactive platform through which a mute person can communicate using sign language or text/visuals.

We envision this product to be utilized in customer-care support for specially abled and kiosks in shopping malls.

Product Al Canvas

Al Canvas

Opportunity

Better user experience for customers with hearing or speaking aid

Users

specially abled customers such as deaf and mute individuals

Strategy

Understanding the sign language and replying in text

Solution

To read user's gestures through their webcam, process the message and interact with them through text.

Data

MNSIT (A-Z signs) ASL, (O-9, A-Z signs) WLASL (Video data)

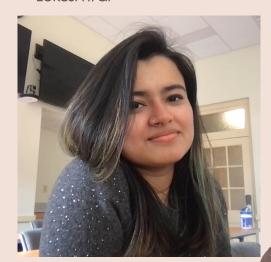
Success Criteria

Correct response to user

Team



Lokeshwar



Afia



Zhiping



Pranav

Value of Data Calculation for User

Improving Customer Experience

Texting is one way to interact, it lacks the personalized touch, especially for companies who value customer satisfaction at their core.

This product addresses the needs of the specially abled community to make them feel more welcomed and engaged.



Data Flywheel To convert Problem Data hand signs Definition Curation to texts MNSIT, ASL, WLASL 2000 Streamlit Mongo DB **TDD Testing** CNN 3 Layer CNN 5 Layer Model **VGG 16** Deployment/ Inception V3/ Model Monitoring Development

Model Type Selection

Logistic Regression

Baseline Model

CNN 3-Layers, CNN 5-Layers

CNNs work well with image classification tasks

Transfer learning with VGG-16

Updated the last layer for our use-case while preserving the pretrained weights



Transfer learning with Inception V3

[Inception blocks after convolving with multiple filters seems to give us the best results by far]

Model Metrics

Accuracy

Model Test Accuracy

Precision

Precision on all the symbols, letters & digits

Recall on all the symbols, letters & digits

Testing in real-time scenarios

[We tried to test the model performances on real-time image captures through streamlit app]

Model Results

Models	Dataset	Accuracy
Logistic Regression	MNIST Dataset (28°28) Alphabet Images (200°200)	Train: 77.5% Test: 74.37% Train: 72% Test: 68.4%
CNN 3 Layers	MNIST Dataset (28°28) Alphabet Images (200°200)	Train: 95.1% Test: 93.2% Train: 98.1% Test: 97.3%
CNN 5 Layers	MNIST Dataset (28°28) Alphabet Images (200°200)	Train: 96.2% Test: 95.3% Train: 98.3% Test: 97.4%
VGG-16	MNIST Dataset (28°28) Alphabet Images (200°200)	Train: 99.8% Test: 99.01% Train: 97.2% Test: 98.5%
Inception V3	MNIST Dataset (28*28) Alphabet Images (200*200)	Train: 99.9% Test: 99.7% Train: 98.6% Test: 98.9%

Inferences

- Two signs, "M" and "S" are being confused by the model.
- Higher resolution images work well
- Pretrained models seem to work well on the datasets we trained on
- For real-time sign to text conversion, model performance worsens from the experiments
- Various data augmentation would help us in the real-time prediction
- Fingerspelling of larger words and sentences is not very suitable with the current model.

MVP development

StreamLIT UI App



User Inputs Image of Sign

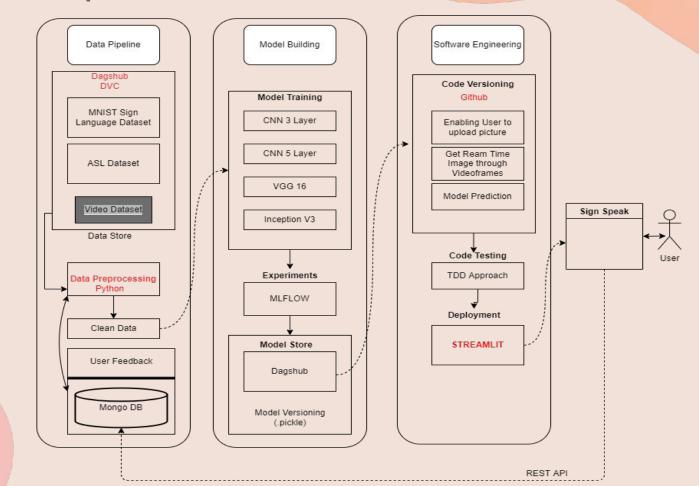


App responds with the meaning/interpretation

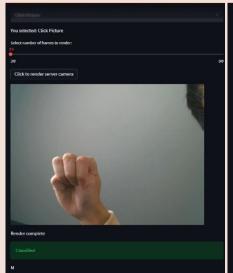


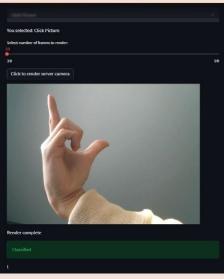
[Product envisioned to be used by customer service orgs/kiosks to better service everyone]

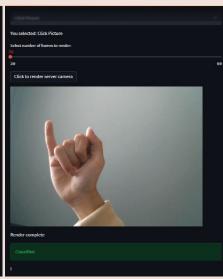
The Proposed Architecture



End-user Application









M

L

Governance & Monitoring

- The dataset that we used is heavily biased on whites.
- Model doesn't seem to work well in other scenarios.
- Planning to scrape balanced data to further improve the model
- Sign languages are often region based. In order to make it more robust, can train for other languages
- Monitoring model and data drift through the user feedback in order to improve model performance.

Project Repo

- Dagshub: https://dagshub.com/loki2124/SignSpeak
- Streamlit: localhost [for now]

THANK YOU!

